# Design, Integration, and Flight Test Results for an Autonomous Surveillance Helicopter

AHS UAV Specialist Meeting January 19, 2005



Dr. M. Takahashi QSS, Computational Sciences Division NASA Ames Research Center, CA

G. Schulein SJS, Army/NASA Rotorcraft Division NASA Ames Research Center, CA M. Whalley Army/NASA Rotorcraft Division Ames Research Center, CA

J. Howlett Army/NASA Rotorcraft Division Ames Research Center, CA Dr. M. Freed Computational Sciences Division NASA Ames Research Center, CA

R. Harris QSS, Computational Sciences Division NASA Ames Research Center, CA

### **Outline**

- Introduction
- Airborne Surveillance Planning Problem
- Evaluating Surveillance Algorithms
- Flight Implementation
- Flight Test Results
- Concluding Remarks

## Introduction Autonomous Rotorcraft Project

 NASA (Intelligent Systems) and US Army AFDD



## Introduction Autonomous Rotorcraft Project

- NASA (Intelligent Systems) and US Army AFDD
- Modeling and simulation



## Introduction Autonomous Rotorcraft Project

- NASA (Intelligent Systems) and US Army AFDD
- Modeling and simulation
- Flight control



- NASA (Intelligent Systems) and US Army AFDD
- Modeling and simulation
- Flight control
- Route planning for obstacle avoidance



- NASA (Intelligent Systems) and US Army AFDD
- Modeling and simulation
- Flight control
- Route planning for obstacle avoidance
- Stereo vision/laser mapping



- NASA (Intelligent Systems) and US Army AFDD
- Modeling and simulation
- Flight control
- Route planning for obstacle avoidance
- Stereo vision/laser mapping
- Autonomous landing at noncooperative sites



- NASA (Intelligent Systems) and US Army AFDD
- Modeling and simulation
- Flight control
- Route planning for obstacle avoidance
- Stereo vision/laser mapping
- Autonomous landing at noncooperative sites
- Contingency planning

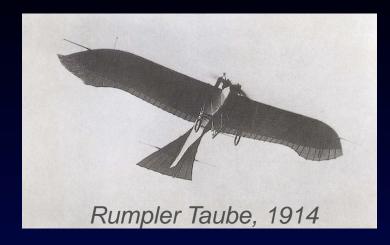


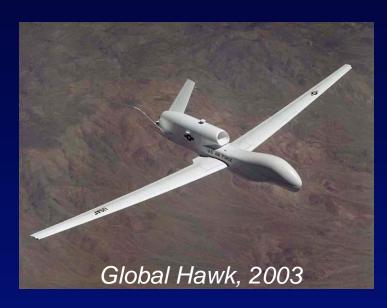
- NASA (Intelligent Systems) and US Army AFDD
- Modeling and simulation
- Flight control
- Route planning for obstacle avoidance
- Stereo vision/laser mapping
- Autonomous landing at noncooperative sites
- Contingency planning
- Autonomous surveillance



### Introduction Surveillance and UAVs

- One of the earliest applications of air vehicles was surveillance
  - Artillery guidance
  - Security
  - Land management
  - Science
- UAVs dramatically increase the availability of surveillance platforms
  - Lower cost
  - More diverse, (possibly) less sophisticated users
- Surveillance is lengthy, repetitive, and largely uneventful making it an ideal candidate for autonomy



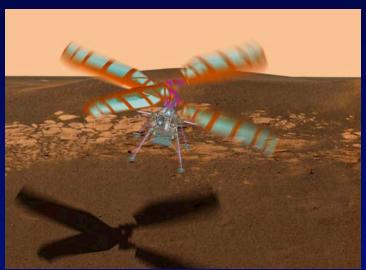


### Introduction Goals

- Develop new autonomous surveillance algorithms
- Implement simulation and flight testbeds for evaluating methods
- Evaluate algorithms/humans to determine which is best for a given situation
- Create a theoretical foundation for surveillance:

Airborne Surveillance Planning Problem



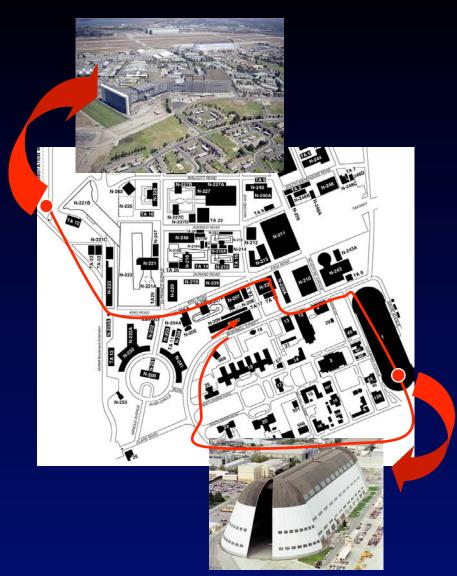


### **Outline**

- Introduction
- Airborne Surveillance Planning Problem
- Evaluating Surveillance Algorithms
- Flight Implementation
- Flight Test Results
- Concluding Remarks

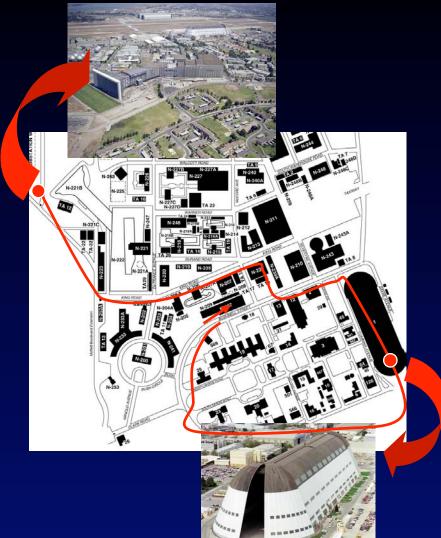
## Airborne Surveillance Planning Problem Problem Description

- Repeated or continuous observations to maintain awareness of entity or geographical area
- Need to decide:
  - where to go next
  - what actions to perform



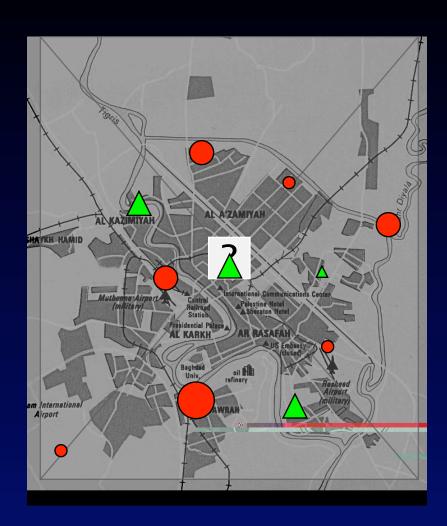
## Airborne Surveillance Planning Problem **Problem Description**

- Diverse site visit requirements
  - Range of importance/value
  - Some require repeated visits
  - Sites might well be omitted
  - Dynamic targets
- Payload effects
  - Sensor type
  - Lighting conditions
  - Limited pointing capability
- Air vehicle effects
  - Travel time obstacle avoidance
  - Environmental effects
- Spatial and temporal effects
  - Uniform distribution vs. "clumping"
  - Sites may suddenly be added or removed
     Unlikely that a single algorithm will cover all cases

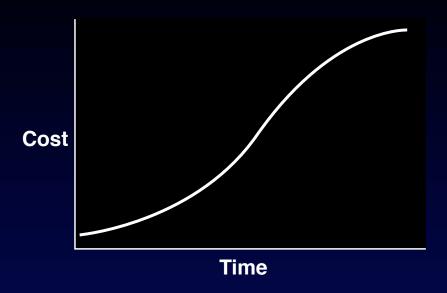


## Airborne Surveillance Planning Problem A Sample Problem

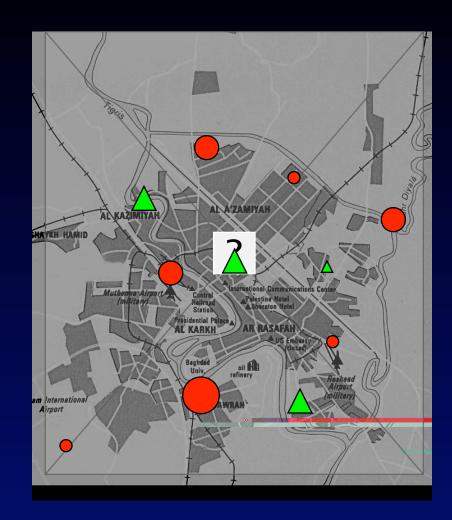
- Valuable Assets
  - perimeter gates
  - warehouses
  - roads
  - airports
- Risk any asset can start on fire, broken into, etc.
- UAV Goal do a good job detecting events thereby mitigating losses

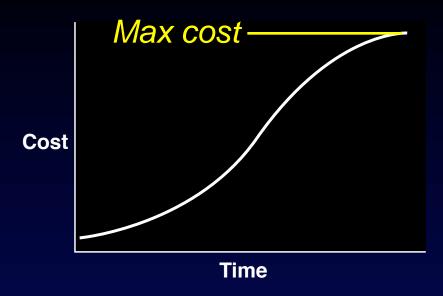


How to quantify "good job"?

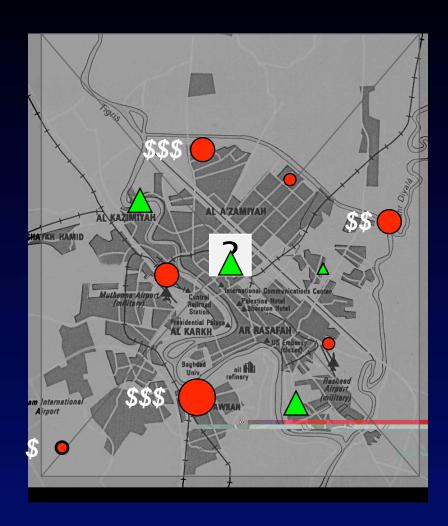


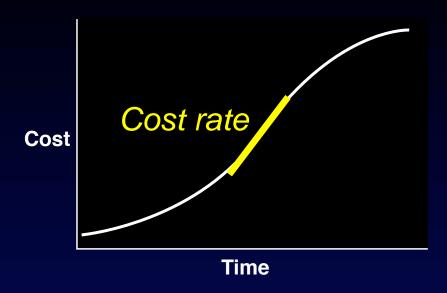
 Each target has an associated cost function (Massios, 2001)



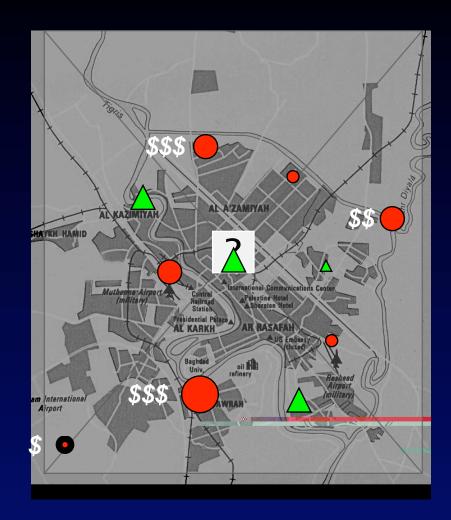


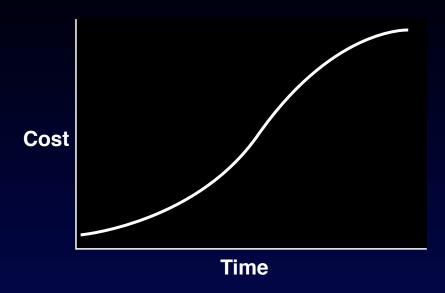
- Each target has an associated cost function (Massios, 2001)
- Some targets are more valuable



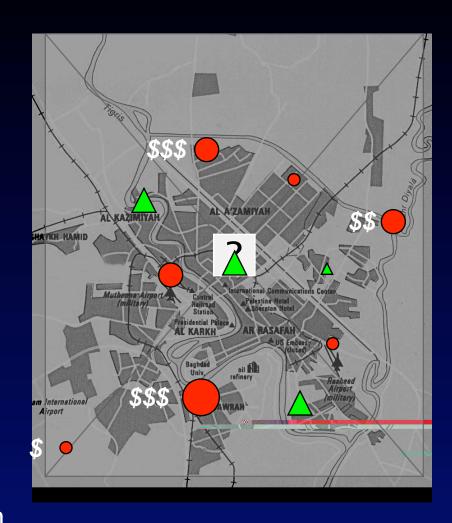


- Each target has an associated cost function (Massios, 2001)
- Some targets are more valuable
- Some targets accrue cost more rapidly





- Each target has an associated cost function (Massios, 2001)
- Some targets are more valuable
- Some targets accrue cost more rapidly
- Probability of occurrence varies with each target



#### Airborne Surveillance Planning Problem Expected Cost of Ignorance

cost of occurrence function (e.g. sigmoid)

$$ECI_{\tau}(t_{1}, t_{2}) = \int_{t=t_{1}}^{t_{2}} p(t) \cos t(t_{2} - t) dt$$
event probability density function (e.g., exponential)

- Flight dynamics and travel time integrated into ECI
- Goal is to minimize ECI

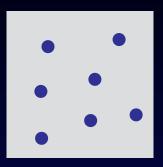
### **Outline**

- Introduction
- Airborne Surveillance Planning Problem
- Evaluating Surveillance Algorithms
- Flight Implementation
- Flight Test Results
- Concluding Remarks

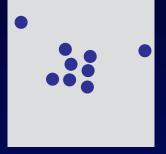
## Evaluating Surveillance Algorithms Defining the Problem Space

- Problem space definition
  - 1. Count
  - 2. Spatial scale
  - 3. Spatial distribution
  - 4. Maximum cost loss
  - 5. Cost rate
- Characterization permits direct comparison of solution methods

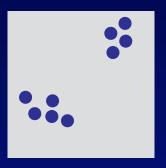
Uniform



Globular



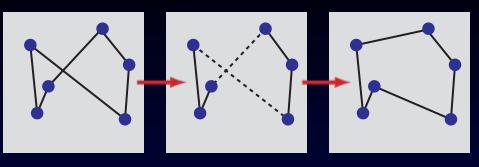
2-Cluster



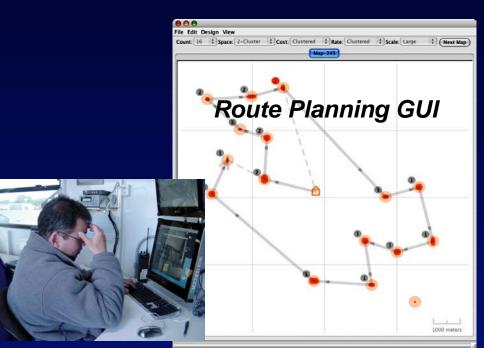
#### **Evaluating Surveillance Algorithms**

#### A Comparative Analysis – 2-Opt versus Human

- GUI symbology indicated max cost, cost rate
- 243 scenarios ~ 6 hours
- Five subjects
- Approximated experienced operators
  - No time limit
  - Training and practice
  - Scoring feedback



2-Opt Solution



Human vs 2-Opt Study

## Evaluating Surveillance Algorithms A Comparative Analysis – 2-Opt versus Human Results

7				4		4 Total	1	8		8 Total	
Scale	Rate	Cost	2-Cluster	Globular	Uniform	Section 2	2-Cluster	Globular	Uniform		2-Ch
Large	Clustered	Clustered	-3			-1	2	-1	-8	-3	-5
		Fixed	-3			-1	-2	-2	-4	-3	2
		Uniform	-4		-1	-1	-3	1	-6	-3	-7
	Clustered Total		-3			-1	-1	-1	-6	-3	-3
	Fixed	Clustered	-3			-1	-2		-6	-3	-8
		Fixed	-3			-1	-2	-1	-3	-2	1
		Uniform	-5		-1	-2	-3	1	-5	-2	-7
		Fixed Total	-4			-1	-2		-4	-2	-5
	Uniform	Clustered	-4	-1	1	-1	-3	-2	-9	-5	-5
		Fixed	-4	-1	1	-1	-2	-3	-2	-2	
		Uniform	-5	-3	-2	-3	-3	2	-7	-3	-9
	U	-4	-2		-2	-3	-1	-6	-3	-5	
Large Total			-4	-1		-2	-2	-1	-6	-3	-4
17		Clustered		100				1	Y.	1	3

### **Outline**

- Introduction
- Airborne Surveillance Planning Problem
- Evaluating Surveillance Algorithms
- Flight Implementation
- Flight Test Results
- Concluding Remarks











## Implementation Avionics Payload

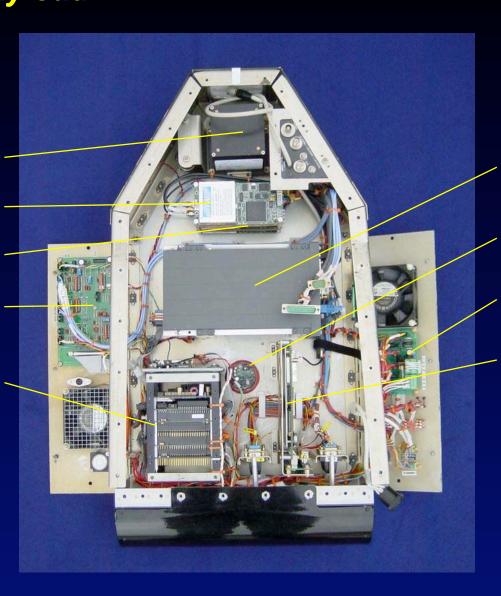
**Crossbow AHRS IMU** 

900 MHz radio modem

servo controller

**Analog conditioning** 

PC104+ flight computer



Compact PCI video computer with 802.11b

Sonar

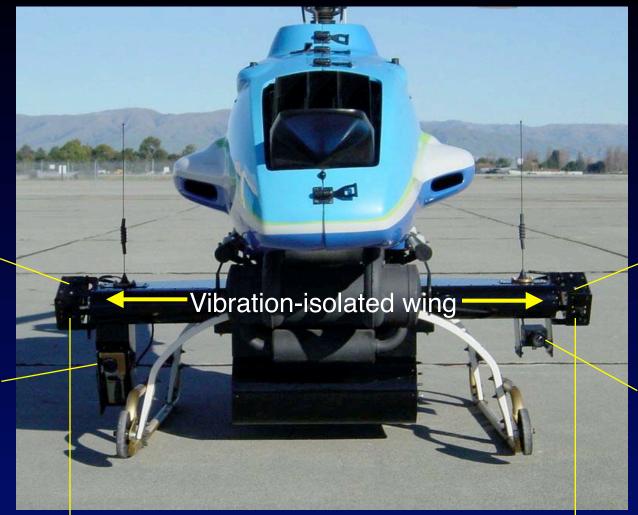
**Power distribution** 

**Ashtech DGPS** 

### **Stub Wing and Cameras**

Firewire camera (mono)

Camcorder



Firewire cameras

Color Unibrain Fire-i400

1.1 meter baseline

### Implementation Stereo Firewire Cameras





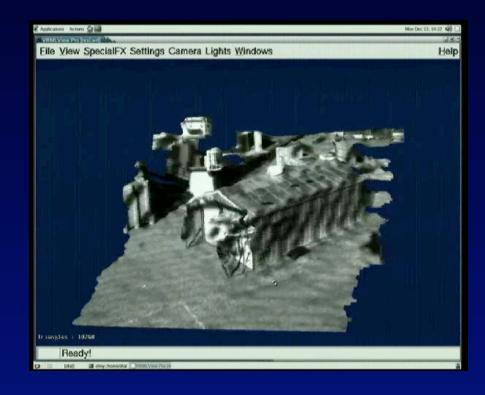
- Point Grey Research Flea, 45g
- Fixed-focal-length lens, 8 mm, 48g
- Stereo tilt system

#### Passive Obstacle Sensing and Mapping

- Images grabbed from server
- Range map generated using Stereo Pipeline
  - Software implementation gives 160 x 120 disparity map once every 5 seconds
- Real-time transformation to inertial coordinates using 6DOF DGPS/IMU solution and camera tilt
- Wide baseline gives accurate disparity at long distances







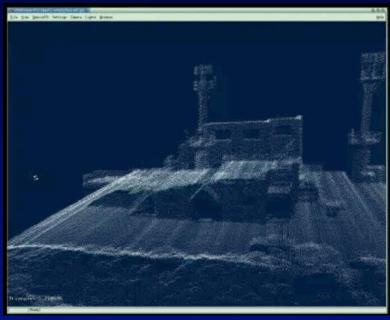
### Active Obstacle Sensing and Mapping





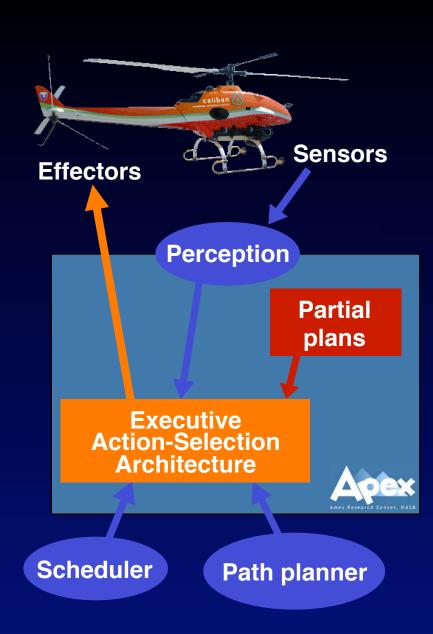
- Weight reduced 9.9 lb to 3.6 lb
- 180 degree field-of-view
- 13,500 points/sec
- 1 deg resolution at 75 Hz
- 0.5 deg resolution at 37.5 Hz
- 81 m range, 1 cm accuracy
- Reposition-able mount





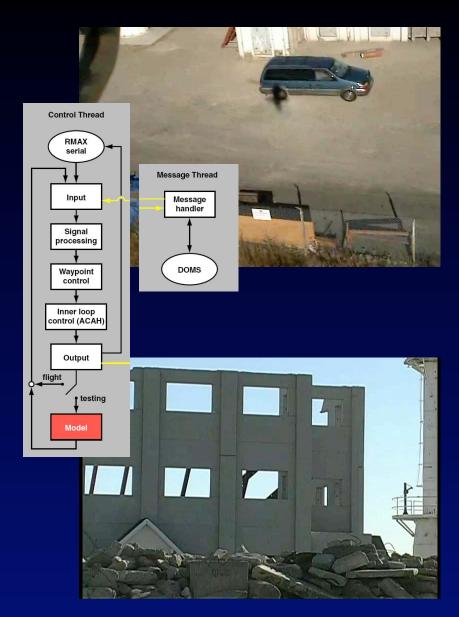
#### **Apex Reactive Planner**

- Apex executive architecture constructs solution using library of partial plans (PDL) containing pre-defined contingencies
- Specialists solve subcomponents of the overall plan
- Provides for creation, simulation, and analysis of agent performance (SHERPA)
- Reduces time required for modeling elemental behavior from which complex plans emerge
- Evolving under this project



## Implementation Flight Control System

- Model-following control law provides attitude stabilization and waypoint guidance
- Path smoothing
  - Kochanek-Bartels cubic spline fit on-the-fly within pre-defined safe corridor
  - Speed profile to respect predefined pitch, bank angle, and climb/descent rate limits
- Control law maintains independent heading control modes



### Desktop and Hardware-in-the-Loop Simulation

- Identical software load used for hardware-in-the-loop testing
- Integrated math model includes:
  - Validated hover/low speed and forward-flight linear models identified from flight data
  - Actuator dynamic models
  - Sensor quantization and noise
  - Transport delays



### **Outline**

- Introduction
- Airborne Surveillance Planning Problem
- Evaluating Surveillance Algorithms
- Flight Implementation
- Flight Test Results
- Concluding Remarks

### Flight Test Results Mission Behavior

- Operator selects targets-of-interest
- Apex computes vantage points
- Apex uses 2-opt to compute sequence
- Obstacle avoidance modifies route as needed
- Targets-of-interest added and deleted



### **Video**

### Concluding Remarks

- Surveillance problem defined
- Expected Cost of Ignorance
- Evaluation methodology
- Human versus modified 2-opt
- Flight demonstration
- Robust, flexible autonomous research helicopter platform



### Future Work - FY05

- NASA IS Final demonstration of full surveillance mission
  - Target sequencing
  - Obstacle detection and mapping
  - Route planning
  - Contingency planning for RF and camera failure
- US Army S&T Landing at non-cooperative site
  - Safe landing area determination



# Design, Integration, and Flight Test Results for an Autonomous Surveillance Helicopter

AHS UAV Specialist Meeting January 19, 2005



Dr. M. Takahashi QSS, Computational Sciences Division NASA Ames Research Center, CA

G. Schulein SJS, Army/NASA Rotorcraft Division NASA Ames Research Center, CA M. Whalley Army/NASA Rotorcraft Division Ames Research Center, CA

J. Howlett Army/NASA Rotorcraft Division Ames Research Center, CA Dr. M. Freed Computational Sciences Division NASA Ames Research Center, CA

R. Harris QSS, Computational Sciences Division NASA Ames Research Center, CA